Conforming and non-conforming functional \textit{a posteriori} error estimates for elliptic boundary value problems in exterior domains: theory and numerical tests

O. MALI*, A. MUZALEVSKIY†, and D. PAULY‡§

Abstract — This paper is concerned with the derivation of conforming and non-conforming functional \textit{a posteriori} error estimates for elliptic boundary value problems in exterior domains. These estimates provide computable and guaranteed upper and lower bounds for the difference between the exact and the approximate solution of the respective problem. We extend the results from [5] to non-conforming approximations, which might not belong to the energy space and are just considered to be square integrable. Moreover, we present some numerical tests.

1. Introduction

Similarly to [5], we consider the standard elliptic Dirichlet boundary value problem

\begin{align*}
-\text{div} \, A \nabla u &= f & \text{in } \Omega \\
 u &= u_0 & \text{on } \Gamma
\end{align*}

where \( \Omega \subset \mathbb{R}^N \) with \( N \geq 3 \) is an exterior domain, i.e., a domain with a compact complement, having for simplicity a Lipschitz continuous boundary

*Department of Mathematical Information Technology, University of Jyväskylä, FI-40014, Finland
†Applied Mathematics Department, St. Petersburg Polytechnic University, St. Petersburg 195251, Russia
‡Faculty für Matematik, Universität Duisburg–Essen, Campus Essen, Thea-Leymann Str. 9, 45141 Essen, Germany
§University of Jyväskylä, FI-40014, Finland
\( \Gamma := \partial \Omega \). Moreover, \( A : \Omega \to \mathbb{R}^{N \times N} \) is a real and symmetric \( L^\infty(\Omega) \)-matrix valued function such that

\[
\exists \alpha > 0 \quad \forall \xi \in \mathbb{R}^N \quad \forall x \in \Omega \quad A(x) \xi \cdot \xi \geq \alpha^{-2} |\xi|^2
\]

holds. As usual, when working with exterior domain problems, we use the polynomially weighted Lebesgue spaces

\[
L^2_s(\Omega) := \{ u : \rho^s u \in L^2(\Omega) \}, \quad \rho := (1 + r^2)^{1/2} \cong r, \quad s \in \mathbb{R}
\]

where \( r(x) := |x| \) is the absolute value. Throughout the paper at hand we just need the values \( s \in \{-1, 0, 1\} \) of the weights. If \( s = 0 \), then we write \( L^2(\Omega) := L^2_0(\Omega) \). Moreover, we introduce the polynomially weighted Sobolev spaces

\[
H^1_{-1}(\Omega) := \{ u : u \in L^2_{-1}(\Omega), \nabla u \in L^2(\Omega) \}
\]

\[
D(\Omega) := \{ v : v \in L^2(\Omega), \text{div} v \in L^2_1(\Omega) \}
\]

which we equip as \( L^2_s(\Omega) \) with the respective scalar products. We will not distinguish in our notation between scalar and vector valued spaces. Moreover, to model homogeneous boundary traces, we define as a closure of test functions

\[
\overset{\circ}{H}^1_{-1}(\Omega) := \overset{\circ}{\overline{H}}^1_{-1}(\Omega) := C^\infty(\Omega).
\]

Note that all these spaces are Hilbert spaces and for the norms we have

\[
|u|^2_{L^2_s(\Omega)} = |\rho^s u|^2_{L^2(\Omega)} = \int_{\Omega} (1 + r^2)^s |u|^2 \, dl
\]

\[
|u|^2_{H^1_{-1}(\Omega)} = |\rho^{-1} u|^2_{L^2(\Omega)} + |\nabla u|^2_{L^2(\Omega)}
\]

\[
|v|^2_{D(\Omega)} = |v|^2_{L^2(\Omega)} + |\rho \text{div} v|^2_{L^2(\Omega)}.
\]

Also, let us introduce for vector fields \( v \in L^2(\Omega) \) the weighted norm

\[
|v|_{L^2(\Omega), A} := \langle v, v \rangle_{L^2(\Omega), A}^{1/2} := \langle Av, v \rangle_{L^2(\Omega)}^{1/2} = |A^{1/2} v|_{L^2(\Omega)}.
\]

(1.3)

Let

\[
c_N := \frac{2}{N-2}, \quad c_{N, \alpha} := \alpha c_N.
\]
From [2] we cite the Poincaré estimate III (see also the appendix of [5]):

\[ \forall u \in \overset{\circ}{H}^1_{-1}(\Omega) \quad |u|_{L^2_{-1}(\Omega)} \leq c_N |\nabla u|_{L^2(\Omega)} \leq c_{N, \alpha} |\nabla u|_{L^2(\Omega), A} \quad (1.4) \]

which is the proper coercivity estimate for the problem at hand. Using this estimate, it is not difficult by standard Lax–Milgram theory of unique solutions to get \( \hat{u} \in \overset{\circ}{H}^1_{-1}(\Omega) + \{u_0\} \) of (1.1)–(1.2) depending continuously on the data for any \( f \in L^2_1(\Omega) \) and \( u_0 \in H^1_{-1}(\Omega) \). Note that the solution \( \hat{u} \) satisfies the variational formulation

\[ \forall u \in \overset{\circ}{H}^1_{-1}(\Omega) \quad \langle \nabla \hat{u}, \nabla u \rangle_{L^2(\Omega), A} = \langle A \nabla \hat{u}, \nabla u \rangle_{L^2_1(\Omega), A} = \langle f, u \rangle_{L^2(\Omega)} \quad (1.5) \]

where we use the \( L^2(\Omega) \)-inner product notation also for the \( L^2_{-1}(\Omega)-L^2_1(\Omega) \)-duality. Note that

\[ \langle f, u \rangle_{L^2(\Omega)} = \int_\Omega f u \, dl \]

is well defined, since the product \( f u \) belongs to \( L^1(\Omega) \). Moreover, we note that

\[ A \nabla \hat{u} \in D(\Omega), \quad -\text{div} A \nabla \hat{u} = f. \]

Let \( \tilde{u} \) be an approximation of \( \hat{u} \). The aim of this contribution is twofold. First, we extend the results from [5] to non-conforming approximations \( \tilde{u} \), which no longer necessarily belong to the natural energy space \( H^1_{-1}(\Omega) \) and hence lack regularity. We will just assume that we have been given an approximation \( \tilde{v} \in L^2(\Omega) \) of \( A \nabla \hat{u} \) without any regularity except of \( L^2(\Omega) \). Second, we validate the conforming \textit{a posteriori} error estimates for problem (1.1)–(1.2) by numerical computations in the exterior domain \( \Omega \) and, therefore, demonstrate that this technique also works in unbounded domains. Such \textit{a posteriori} error estimates have been extensively derived and discussed earlier for problems in bounded domains (see, e.g., [4, 7, 8]) and the literature cited therein. The underlying general idea is to construct estimates via Lagrangians. In linear problems this can be done by splitting the residual functional into two natural parts using simply integration by parts of relations, which then immediately yields guaranteed and computable lower and upper bounds. In fact, one adds a zero to the weak form.
2. Conforming \textit{a posteriori} estimates

For the convenience of the reader, we repeat for the conforming case the main arguments from [5] to obtain the desired \textit{a posteriori} estimates. We want to deduce estimates for the error
\[ e := \hat{u} - \tilde{u} \]
in the natural energy norm $|\nabla e|_{L^2(\Omega), A}$. In this section we only consider conforming approximations, i.e., $\tilde{u} \in \tilde{H}^1_{-1}(\Omega) + \{u_0\}$ and therefore $e \in \tilde{H}^1_{-1}(\Omega)$.

Introducing an arbitrary vector field $v \in D(\Omega)$ and inserting a zero into (1.5), for all $u \in \tilde{H}^1_{-1}(\Omega)$ we have
\[
\langle \nabla (\hat{u} - \tilde{u}), \nabla u \rangle_{L^2(\Omega), A} = \langle f, u \rangle_{L^2(\Omega)} - \langle A\nabla \tilde{u} - v, \nabla u \rangle_{L^2(\Omega)} \\
= \langle f + \text{div} v, u \rangle_{L^2(\Omega)} - \langle \nabla \tilde{u} - A^{-1}v, \nabla u \rangle_{L^2(\Omega), A}
\]
(2.1)
since $u$ satisfies the homogeneous boundary condition. Therefore, by (1.4) we have
\[
\langle \nabla e, \nabla u \rangle_{L^2(\Omega), A} \leq |f + \text{div} v|_{L^2(\Omega)}^2 |u|_{L^2(\Omega)}^2 + |\nabla \tilde{u} - A^{-1}v|_{L^2(\Omega), A}^2 |\nabla u|_{L^2(\Omega), A}^2 \\
\leq \left( c_{N, \alpha} |f + \text{div} v|_{L^2(\Omega)}^2 + |\nabla \tilde{u} - A^{-1}v|_{L^2(\Omega), A}^2 \right) |\nabla u|_{L^2(\Omega), A}^2 \\
=: M_+(\nabla \tilde{u}, v; f, A) =: M_+(\nabla \tilde{u}, v)
\]
(2.2)
Taking $u := e \in \tilde{H}^1_{-1}(\Omega)$ we obtain the upper bound
\[
|\nabla e|_{L^2(\Omega), A} \leq \inf_{v \in D(\Omega)} M_+(\nabla \tilde{u}, v).
\]
(2.3)
We note that for $v := A\nabla \hat{u}$ we have $M_+(\nabla \tilde{u}, v) = |\nabla e|_{L^2(\Omega), A}$. Therefore, in (2.3) we even have equality.

The lower bound can be obtained as follows. Let $u \in \tilde{H}^1_{-1}(\Omega)$. Then, by the trivial inequality $|\nabla (\hat{u} - \tilde{u}) - \nabla u|_{L^2(\Omega), A} \geq 0$ and (1.5) we obtain
\[
|\nabla (\hat{u} - \tilde{u})|_{L^2(\Omega), A}^2 \geq 2 |\nabla (\hat{u} - \tilde{u}), \nabla u \rangle_{L^2(\Omega), A} - |\nabla u|_{L^2(\Omega), A}^2 \\
= 2 \langle f, u \rangle_{L^2(\Omega)} - \langle \nabla (2\tilde{u} + u), \nabla u \rangle_{L^2(\Omega), A} \\
=: M_-(\nabla \tilde{u}, u; f, A) =: M_-(\nabla \tilde{u}, u)
\]
and hence we get the lower bound

\[ |\nabla e|_{L^2(\Omega),A}^2 \geq \sup_{u \in \hat{H}^1_1(\Omega)} M_-(\nabla \hat{u}, u). \] (2.4)

Again, we note that for \( u := \hat{u} - \tilde{u} = e \) we have \( M_-(\nabla \tilde{u}, u) = |\nabla e|_{L^2(\Omega),A}^2 \). Therefore, also in (2.4) the equality holds. Let us summarize the result.

**Theorem 2.1 (conforming a posteriori estimates).** Let \( \tilde{u} \in \hat{H}^1_1(\Omega) + \{u_0\} \). Then

\[ \max_{u \in \hat{H}^1_1(\Omega)} M_-(\nabla \tilde{u}, u) = |\nabla (\hat{u} - \tilde{u})|_{L^2(\Omega),A}^2 = \min_{v \in D(\Omega)} M_+^2(\nabla \tilde{u}, v) \]

where the upper and lower bounds are given by

\[
M_+(\nabla \tilde{u}, v) = c_{N,\alpha} |f + \text{div} v|_{L^2_1(\Omega)}^2 + |\nabla \tilde{u} - A^{-1}v|_{L^2(\Omega),A}^2 \\
M_-(\nabla \tilde{u}, u) = 2 \langle f, u \rangle_{L^2(\Omega)} - \langle \nabla (2\tilde{u} + u), \nabla u \rangle_{L^2(\Omega),A}.
\]

The functional error estimators \( M_+ \) and \( M_- \) are referred as the majorant and the minorant, respectively. They possess the usual features, e.g., they contain just one constant \( c_{N,\alpha} \), which is well known, and they are sharp. Hence the variational problems for \( M_\pm \) themselves provide new and equivalent variational formulations for system (1.1)–(1.2). Moreover, since the functions \( u \) and vector fields \( v \) are at our disposal, one can generate different numerical schemes to estimate the energy norm of the error \( |\nabla e|_{L^2(\Omega),A} \). For details see, e.g., [3, 4, 8] and references therein.

**Remark 2.1.** It is often desirable to have the majorant in the quadratic form

\[ M_+^2(\nabla \tilde{u}, v) \leq \inf_{\beta > 0} \left( c_{N,\alpha}^2 (1 + 1/\beta) |f + \text{div} v|_{L^2_1(\Omega)}^2 + (1 + \beta) |\nabla \tilde{u} - A^{-1}v|_{L^2(\Omega),A}^2 \right). \]

This form is well suited for computations, since the minimization with respect to the vector fields \( v \in D(\Omega) \) over some finite dimensional subspace (e.g., generated by finite elements) reduces to solving a system of linear equations.
Remark 2.2. If $v \approx A \nabla \hat{u}$, then the first term of the majorant is close to zero and the second term can be used as an error indicator to study the distribution of the error over the domain, i.e.,

$$(\nabla \bar{u} - A^{-1}v) \cdot (A \nabla \bar{u} - v) \approx \nabla (\bar{u} - \hat{u}) \cdot A \nabla (\bar{u} - \hat{u})$$

in $\Omega$.

The question how to measure the actual performance of the error indicator (the ‘accuracy’ of the symbol $\approx$) is addressed extensively in the forthcoming book [3].

3. Non-conforming a posteriori estimates

To achieve estimates for non-conforming approximations $\bar{u} \notin \overset{\circ}{H}^{-1}_1(\Omega) + \{u_0\}$, we utilize the simple Helmholtz decomposition

$$L^2(\Omega) = \nabla \overset{\circ}{H}^{-1}_1(\Omega) \oplus_A A^{-1}D_0(\Omega),$$  \hspace{1cm} (3.1)

where $D_0(\Omega) := \{v \in D(\Omega) : \text{div} v = 0\}$ and $\oplus_A$ denotes the orthogonal sum with respect to the weighted $A$-$L^2(\Omega)$-scalar product, see (1.3). The decomposition (3.1) follows immediately from the projection theorem in Hilbert spaces

$$L^2(\Omega) = \nabla \overset{\circ}{H}^1(\Omega) \oplus_A A^{-1}D_0(\Omega)$$

and the fact that $\nabla \overset{\circ}{H}^{-1}_1(\Omega) = \nabla \overset{\circ}{H}^1(\Omega)$ is closed in $L^2(\Omega)$ by (1.4). Note that the negative divergence

$$- \text{div} : D(\Omega) \subset L^2(\Omega) \to L^2(\Omega)$$

is the adjoint of the gradient

$$\overset{\circ}{\nabla} : H^1(\Omega) \subset L^2(\Omega) \to L^2(\Omega)$$

i.e., $\overset{\circ}{\nabla}^* = - \text{div}$, and thus $L^2(\Omega) = \overset{\circ}{R}(\overset{\circ}{\nabla}) \oplus N(\overset{\circ}{\nabla}^*)$. Here, we have used the unweighted standard Sobolev spaces $H^1(\Omega)$ and $D(\Omega)$ and the notation $R$ and $N$ for the range and the null space or the kernel of a linear operator, respectively.
Let us now assume that we have an approximation \( \tilde{v} \in L^2(\Omega) \) of \( A\nabla \tilde{u} \). According to (3.1), we decompose the ‘gradient-error’ orthogonally

\[
\mathcal{L}^2(\Omega) \ni E := \nabla \tilde{u} - A^{-1} \tilde{v} = \nabla \varphi + A^{-1} \psi, \quad \varphi \in \mathcal{H}^1(\Omega), \quad \psi \in D_0(\Omega)
\]  

(3.2)

and note that it decomposes by Pythagoras’ theorem into

\[
|E|_{L^2(\Omega), A}^2 = |\nabla \varphi|_{L^2(\Omega), A}^2 + |A^{-1} \psi|_{L^2(\Omega), A}^2
\]  

(3.3)

which allows us to estimate the two error terms separately.

For \( u \in \mathcal{H}^1(\Omega) \) we have by orthogonality

\[
\langle \nabla \varphi, \nabla u \rangle_{L^2(\Omega), A} = \langle E, \nabla u \rangle_{L^2(\Omega), A} = \langle f, u \rangle_{L^2(\Omega)} - \langle \tilde{v}, \nabla u \rangle_{L^2(\Omega)}.
\]

Now we can proceed exactly as in (2.1) and (2.2) replacing \( A\nabla \tilde{u} \) by \( \tilde{v} \). More precisely, for all \( v \in D(\Omega) \) we have

\[
\langle \nabla \varphi, \nabla u \rangle_{L^2(\Omega), A} = \langle f, u \rangle_{L^2(\Omega)} - \langle \tilde{v} - v + v, \nabla u \rangle_{L^2(\Omega)}
\]

\[
= \langle f + \text{div} v, u \rangle_{L^2(\Omega)} - \langle \tilde{v} - v, \nabla u \rangle_{L^2(\Omega)}
\]

and hence

\[
\langle \nabla \varphi, \nabla u \rangle_{L^2(\Omega), A} \leq \| f + \text{div} v \|_{L^2(\Omega)} \| u \|_{L^2(\Omega)} + \| A^{-1} (\tilde{v} - v) \|_{L^2(\Omega), A} \| \nabla u \|_{L^2(\Omega), A}
\]

\[
\leq \left( c_{N, A} \| f + \text{div} v \|_{L^2(\Omega)} + \| A^{-1} (\tilde{v} - v) \|_{L^2(\Omega), A} \right) \| \nabla u \|_{L^2(\Omega), A}.
\]

\[
=: M_+ (A^{-1} \tilde{v}, v; f, A) =: M_+ (A^{-1} \tilde{v}, v)
\]

Setting \( u := \varphi \) yields

\[
|\nabla \varphi|_{L^2(\Omega), A} \leq \inf_{v \in D(\Omega)} M_+ (A^{-1} \tilde{v}, v).
\]  

(3.4)

This estimate is no longer sharp, contrary to the conforming case. We just have the equality \( M_+ (A^{-1} \tilde{v}, v) = |E|_{L^2(\Omega), A} \) for \( v = A\nabla \tilde{u} \).

For \( v \in D_0(\Omega) \) and \( u \in \mathcal{H}^1(\Omega) + \{ u_0 \} \), we have by orthogonality and
since $\hat{u} - u$ belongs to $\overset{\circ}{H}^1_{-1}(\Omega)$:

$$\langle A^{-1}\psi, A^{-1}v \rangle_{L^2(\Omega),A} = \langle E, A^{-1}v \rangle_{L^2(\Omega),A}$$

$$= \langle \nabla(\hat{u} - u), v \rangle_{L^2(\Omega)} + \langle \nabla u - A^{-1}\hat{v}, A^{-1}v \rangle_{L^2(\Omega),A}$$

$$= 0$$

$$\leq \left| \nabla u - A^{-1}\hat{v} \right|_{L^2(\Omega),A} L_2(\Omega) A^{-1}v \right|_{L^2(\Omega),A} =: \tilde{M}_+(A^{-1}\hat{v}, \nabla u)$$

Setting $v := \psi$ yields

$$\left| A^{-1}\psi \right|_{L^2(\Omega),A} \leq \inf_{u \in \overset{\circ}{H}^1_{-1}(\Omega) + \{u_0\}} \tilde{M}_+(A^{-1}\hat{v}, \nabla u). \quad (3.5)$$

Again, this estimate is no longer sharp. We just have $\tilde{M}_+(A^{-1}\hat{v}, \nabla u) = |E|_{L^2(\Omega),A}$ for $u = \hat{u}$.

For the lower bounds we pick an arbitrary $u \in \overset{\circ}{H}^1_{-1}(\Omega)$ and compute

$$\left| \nabla \varphi \right|_{L^2(\Omega),A}^2 \geq 2 \langle \nabla \varphi, \nabla u \rangle_{L^2(\Omega),A} - \left| \nabla u \right|_{L^2(\Omega),A}^2$$

$$= \langle E, \nabla u \rangle_{L^2(\Omega),A}$$

$$= 2 \langle \nabla \hat{u}, \nabla u \rangle_{L^2(\Omega),A} - 2 \langle A^{-1}\hat{v}, \nabla u \rangle_{L^2(\Omega),A} - \left| \nabla u \right|_{L^2(\Omega),A}^2$$

$$= 2 \langle f, u \rangle_{L^2(\Omega)} - \langle \nabla u + 2A^{-1}\hat{v}, \nabla u \rangle_{L^2(\Omega),A}$$

$$= M_-(A^{-1}\hat{v}, u; f, A) = M_-(A^{-1}\hat{v}, u).$$

Substituting $u = \varphi$ shows that this lower bound is sharp, since $M_-(A^{-1}\hat{v}, u) = |\nabla \varphi|_{L^2(\Omega),A}^2$.

Now we choose $v \in D_0(\Omega)$ and $u \in \overset{\circ}{H}^1_{-1}(\Omega) + \{u_0\}$ getting

$$\left| A^{-1}\psi \right|_{L^2(\Omega),A}^2 \geq 2 \langle A^{-1}\psi, A^{-1}v \rangle_{L^2(\Omega),A} - \left| A^{-1}v \right|_{L^2(\Omega),A}^2$$

$$= \langle E, A^{-1}v \rangle_{L^2(\Omega),A}$$

$$= 2 \langle \nabla(\hat{u} - u), A^{-1}v \rangle_{L^2(\Omega),A} + 2 \langle \nabla u - A^{-1}\hat{v}, A^{-1}v \rangle_{L^2(\Omega),A}$$

$$= 0$$

$$- \left| A^{-1}v \right|_{L^2(\Omega),A}^2 \leq 2 \langle \nabla u - A^{-1}(2\hat{v} + v), A^{-1}v \rangle_{L^2(\Omega),A}$$

$$= : \tilde{M}_-(A^{-1}\hat{v}, A^{-1}v, \nabla u; A) = : \tilde{M}_-(A^{-1}\hat{v}, A^{-1}v, \nabla u)$$
since $\hat{u} - u \in H^1_0(\Omega)$. Also, this second lower bound is still sharp, since we have $
abla \hat{u} - A^{-1}\tilde{v} \leq |\nabla \phi|_{L^2(\Omega),A}$ for $v = \phi$ and any $u \in H^1_0(\Omega) + \{u_0\}$.

**Theorem 3.1 (non-conforming a posteriori estimates).** Let $\tilde{v} \in L^2(\Omega)$. Then

$$|\nabla \hat{u} - A^{-1}\tilde{v}|^2_{L^2(\Omega),A} \leq M_+(\tilde{v}) := \inf_{v \in D(\Omega)} M_+(A^{-1}\tilde{v}, v)$$

$$+ \inf_{\tilde{u} \in \tilde{H}^1_0(\Omega) + \{u_0\}} \tilde{M}_+(A^{-1}\tilde{v}, \nabla u)$$

$$|\nabla \hat{u} - A^{-1}\tilde{v}|^2_{L^2(\Omega),A} \geq M_-(\tilde{v}) := \sup_{u \in \tilde{H}^1_0(\Omega)} M_-(A^{-1}\tilde{v}, u)$$

$$+ \sup_{u \in \tilde{H}^1_0(\Omega) + \{u_0\}} \sup_{v \in D(\Omega)} \tilde{M}_-(A^{-1}\tilde{v}, A^{-1}v, \nabla u)$$

where

$$M_+(A^{-1}\tilde{v}, v) = c_{N,\alpha} |f + \nabla v|_{L^2(\Omega)} + |A^{-1}(\tilde{v} - v)|_{L^2(\Omega),A}$$

$$\tilde{M}_+(A^{-1}\tilde{v}, \nabla u) = |\nabla u - A^{-1}\tilde{v}|_{L^2(\Omega),A}$$

$$M_-(A^{-1}\tilde{v}, u) = 2 \langle f, u \rangle_{L^2(\Omega)} - \langle \nabla u + \nabla A^{-1}\tilde{v}, \nabla u \rangle_{L^2(\Omega),A}$$

$$\tilde{M}_-(A^{-1}\tilde{v}, A^{-1}v, \nabla u) = \langle 2\nabla u - A^{-1}(2\tilde{v} + v), A^{-1}v \rangle_{L^2(\Omega),A}.$$

Moreover, as in Remark 2.1 we have

$$M_+^2(A^{-1}\tilde{v}, v) \leq \inf_{\beta > 0} \beta \left( c_{N,\alpha}^2 (1 + 1/\beta) |f + \nabla v|_{L^2(\Omega)}^2 \right.$$

$$+ (1 + \beta) |A^{-1}(\tilde{v} - v)|_{L^2(\Omega),A}^2 \left. \right).$$

**Remark 3.1.** The lower bound is still sharp also in this non-conforming estimate. As shown before, taking $u = \varphi$ yields $M_-(A^{-1}\tilde{v}, u) = |\nabla \phi|_{L^2(\Omega),A}$ and for $v = \psi$ and arbitrary $u \in \tilde{H}^1_0(\Omega) + \{u_0\}$ we have

$$\tilde{M}_-(A^{-1}\tilde{v}, A^{-1}v, \nabla u) = |A^{-1}\psi|_{L^2(\Omega),A}^2.$$
Thus, we even have
\[
\left| \nabla \hat{u} - A^{-1} \tilde{v} \right|_{L^2(\Omega),A}^2 \geq \mathcal{M}_-(\tilde{v}) \geq \left| \nabla \varphi \right|_{L^2(\Omega),A}^2 + |A^{-1} \psi|_{L^2(\Omega),A}^2
\]
\[
= \left| \nabla \hat{u} - A^{-1} \tilde{v} \right|_{L^2(\Omega),A}^2
\]
i.e., \( \mathcal{M}_-(\tilde{v}) = \left| \nabla \hat{u} - A^{-1} \tilde{v} \right|_{L^2(\Omega),A}^2 \). The upper bound might no longer be sharp. Taking, e.g., \( u := \tilde{u} \) and \( v = A\nabla \hat{u} \), we get
\[
M_+(A^{-1} \tilde{v}, \nabla u) = \tilde{M}_+(A^{-1} \tilde{v}, \nabla u) = \left| \nabla \hat{u} - A^{-1} \tilde{v} \right|_{L^2(\Omega),A}
\]
and thus \( \mathcal{M}_+(\tilde{v}) \leq 2 \left| \nabla \hat{u} - A^{-1} \tilde{v} \right|_{L^2(\Omega),A}^2 \). So, an overestimation by 2 is possible.

**Remark 3.2.** For conforming approximations \( \tilde{v} = A\nabla \hat{u} \), i.e., \( A^{-1} \tilde{v} = \nabla \hat{u} \), with some \( \tilde{u} \in \tilde{H}^1_0(\Omega) + \{ u_0 \} \), we obtain the estimates from Theorem 2.1, since
\[
\inf_{u \in \tilde{H}^1_0(\Omega)+\{u_0\}} \tilde{M}_+(A^{-1} \tilde{v}, \nabla u) = \inf_{u \in \hat{H}^1_0(\Omega)+\{u_0\}} |\nabla (u - \tilde{u})|_{L^2(\Omega),A} = 0
\]
and
\[
\sup_{u \in \hat{H}^1_0(\Omega)+\{u_0\}} \sup_{v \in D_0(\Omega)} \tilde{M}_-(A^{-1} \tilde{v}, A^{-1} v, \nabla u) = \left\langle 2 \nabla (u - \tilde{u}) - A^{-1} v, A^{-1} v \right\rangle_{L^2(\Omega),A}
\]
\[
= \sup_{v \in D_0(\Omega)} - |A^{-1} v|_{L^2(\Omega),A}^2 = 0
\]
because \( \left\langle \nabla (u - \tilde{u}), A^{-1} v \right\rangle_{L^2(\Omega),A} = \left\langle \nabla (u - \tilde{u}), v \right\rangle_{L^2(\Omega)} = 0 \) by \( u - \tilde{u} \in \hat{H}^1_0(\Omega) \).

**Remark 3.3.** The terms \( \tilde{M}_\pm \) measure the boundary error. To see this, let us introduce the scalar trace operator \( \gamma : H^1_{\Gamma}(\Omega) \to H^{1/2}(\Gamma) \) and a corresponding extension operator \( \tilde{\gamma} : H^{1/2}(\Gamma) \to H^1(\Omega) \). These are both linear and continuous (let’s say with constants \( c_\gamma \) and \( c_{\tilde{\gamma}} \)) and \( \gamma \) is surjective. Moreover, \( \tilde{\gamma} \) is a right inverse to \( \gamma \). For an approximation \( \tilde{v} = A\nabla \hat{u} \) with \( \tilde{u} \in \tilde{H}^1_0(\Omega) \) we
A posteriori error estimates

define \( \tilde{u} := \tilde{u} + \tilde{\gamma}(u_0 - \tilde{u}) \). Then \( \gamma \tilde{u} = \gamma u_0 \) and hence \( \tilde{u} - u_0 \in H_{1}^{1}(\Omega) \). We obtain

\[
\inf_{u \in H_{1}^{1}(\Omega) + \{u_0\}} \frac{\tilde{M}_+(A^{-1}\tilde{v}, \nabla u)}{|\nabla (\tilde{u} - \tilde{u})|_{L^2(\Omega), A} = |\nabla \tilde{\gamma}(u_0 - \tilde{u})|_{L^2(\Omega), A}} \leq c_{\tilde{\gamma}} |\gamma(u_0 - \tilde{u})|_{H^{1/2}(\Gamma)}
\]

and since \( v \in D_0(\Omega) \) by partial integration using the normal trace \( \gamma_{\nu} v \in H^{-1/2}(\Gamma) \) we have

\[
\sup_{u \in H_{1}^{-1}(\Omega) + \{u_0\}} \sup_{v \in D_0(\Omega)} \frac{\tilde{M}_-(A^{-1}\tilde{v}, A^{-1}v, \nabla u)}{2 \langle \nabla (\tilde{u} - \tilde{u}), v \rangle_{L^2(\Omega)} - |A^{-1}v|_{L^2(\Omega), A}^2} = \sup_{v \in D_0(\Omega)} \left(2 \langle \gamma(u_0 - \tilde{u}), \gamma_{\nu} v \rangle_{H^{1/2}(\Gamma), H^{-1/2}(\Gamma)} - |A^{-1}v|_{L^2(\Omega), A}^2\right).
\]

**Remark 3.4.** The results of this contribution can be easily extended in a canonical way to exterior elliptic boundary value problems with pure Neumann or mixed boundary conditions, such as

\[
-\text{div} A \nabla u = f \quad \text{in} \ \Omega \\
\quad u = u_0 \quad \text{on} \ \Gamma_1 \\
\quad \nu \cdot A \nabla u = v_0 \quad \text{on} \ \Gamma_2
\]

where the boundary \( \Gamma \) decomposes into two parts \( \Gamma_1 \) and \( \Gamma_2 \).

4. Numerical tests

Let \( B_R := \{x \in \mathbb{R}^N: |x| < R\} \) denote an open ball, \( E_R := \{x \in \mathbb{R}^N: |x| > R\} \) the exterior domain and \( S_R := \{x \in \mathbb{R}^N: |x| = R\} \) a sphere of the radius \( R \) centered at the origin, respectively, as well as \( \Omega_R := \Omega \cap B_R \). Moreover, let \( N = 3 \), thus \( c_N = 2 \). In our examples we set \( A = \text{id}, f = 0 \) and \( u_0 = 1 \). Hence we have \( \alpha = 1 \) and \( c_{N, \alpha} = 2 \).
Therefore, we will consider the exterior Dirichlet Laplace problem, i.e., find \( \hat{u} \in H^1_{-1}(\Omega) \) such that

\[
\Delta \hat{u} = 0 \quad \text{in } \Omega \tag{4.1}
\]
\[
\hat{u} = 1 \quad \text{on } \Gamma. \tag{4.2}
\]

It is classical that any solution \( u \in L^2_{-1}(\Omega) \) (even \( u \in L^2_s(\Omega) \) with some \( s \geq -3/2 \) is sufficient) of \( \Delta u = 0 \) in \( E_R \) can be represented as as a spherical harmonics expansion with only negative powers, more precisely as a series of spherical harmonics \( Y_{n,m} \) of the order \( n \) multiplied by the proper powers of the radius \( r^{-(n+1)} \), i.e., for \( r > R \):

\[
u_{\Phi}(r, \theta, \varphi) = \sum_{n \geq 0} \sum_{-n \leq m \leq n} \zeta_{n,m} r^{-n-1} Y_{n,m}(\theta, \varphi), \quad \zeta_{n,m} \in \mathbb{R} \tag{4.3}
\]

where \( u_{\Phi} := u \circ \Phi \) and \( \Phi \) denotes the usual polar coordinates (see, e.g., [1, 9]).

**Remark 4.1.** For \( \Omega = E_1 \) the unique solution of (4.1)-(4.2) is \( \hat{u} = 1/r \), which is the first term in expansion (4.3) corresponding to \( n = 0 \). We note that even \( \hat{u} \in L^2_s(\Omega) \), as well as \( |\nabla \hat{u}| = 1/r^2 \in L^2_{s+1}(\Omega) \) hold for every \( s < -1/2 \). For any \( 1 < R < R' \) it is also the unique solution of the exterior Dirichlet Laplace problem

\[
\Delta \hat{u} = 0 \quad \text{in } E_R \\
\hat{u} = 1/R \quad \text{on } S_R
\]

with \( \hat{u} \in H^1_{-1}(E_R) \) and of the Dirichlet Laplace problem

\[
\Delta \hat{u} = 0 \quad \text{in } B_{R'} \cap E_R \\
\hat{u} = 1/R \quad \text{on } S_R \\
\hat{u} = 1/R' \quad \text{on } S_{R'}
\]

with \( \hat{u} \in H^1(B_{R'} \cap E_R) \).

Of course, system (4.1)–(4.2) is equivalent to find \( \hat{u} \in H^1_{-1}(\Omega) \) with \( \hat{u}|_{\Gamma} = 1 \) and

\[
\forall \varphi \in H^1_{-1}(\Omega) \quad \langle \nabla \hat{u}, \nabla \varphi \rangle_{L^2(\Omega)} = 0
\]
or to minimize the energy

\[ \mathcal{E}(u) := |\nabla u|_{L^2(\Omega)}^2 \]

over the set \( \{u \in H^{1-1}(\Omega) : u|_\Gamma = 1\} \). In order to generate an approximate solution \( \tilde{u} \), we split \( \Omega \) into an unbounded and a bounded subdomain, namely \( E_R \) and \( \Omega_R \), where we pick \( R > 0 \) such that \( \mathbb{R}^3 \setminus \Omega \subset B_R \). The domains are depicted in Fig. 1. Our approximation method is based on the assumption that if \( R \) is ‘large enough’, then the first term of the expansion (4.3) dominates in the unbounded subdomain \( E_R \) and hence we simply assume from our approximation \( \tilde{u} \in H^{1-1}(\Omega) \) the asymptotic behaviour

\[ \tilde{u}|_{E_R} = \frac{\zeta}{r} \]

where \( \zeta \) is an unknown real constant. In the bounded subdomain \( \Omega_R \), the approximation \( \tilde{u}_R := \tilde{u}|_{\Omega_R} \in H^1(\Omega_R) \) must satisfy the boundary condition \( \tilde{u}_R|_\Gamma = 1 \) and the continuity condition \( \tilde{u}_R|_{S_R} = \zeta / R \) to ensure \( \tilde{u} \in H^{1-1}(\Omega) \). Then, for approximations of the prescribed type, the problem is reduced to minimize the energy

\[ \mathcal{E}(\tilde{u}_R, \zeta, \zeta) := \mathcal{E}(\tilde{u}) = |\nabla \tilde{u}|_{L^2(\Omega)}^2 = |\nabla \tilde{u}_R, \zeta|_{L^2(\Omega_R)}^2 + \frac{4\pi \zeta^2}{R} \]

with respect to \( \zeta \in \mathbb{R} \) and \( \tilde{u}_R, \zeta \in H^1(\Omega_R) \) with \( \tilde{u}_R, \zeta|_\Gamma = 1 \) and \( \tilde{u}_R, \zeta|_{S_R} = \zeta / R \). We propose an iteration procedure to minimize the quadratic energy

![Figure 1. Domains.](image-url)
or functional $\mathcal{E}$, which is described in Algorithm 4.1 below. It is based on the decomposition

$\tilde{u}_{R,\zeta} = \tilde{u}_{R,\zeta,0} + \tilde{u}_{R,1} + \zeta \tilde{u}_{R,2}$

where $\tilde{u}_{R,\zeta,0} \in \overset{\circ}{H}^1(\Omega_R)$ and $\tilde{u}_{R,1}, \tilde{u}_{R,2} \in H^1(\Omega_R)$ satisfy certain boundary conditions, i.e.,

$\tilde{u}_{R,1}|_{\Gamma} = 1, \quad \tilde{u}_{R,1}|_{S_R} = 0, \quad \tilde{u}_{R,2}|_{\Gamma} = 0, \quad \tilde{u}_{R,2}|_{S_R} = 1 \quad (4.4)$

as well as

$\Delta \tilde{u}_{R,2} \neq 0. \quad (4.5)$

The two functions $\tilde{u}_{R,1}, \tilde{u}_{R,2}$ take care only of the boundary conditions and are fixed during the iteration procedure. This means that the energy $\mathcal{E}(\tilde{u}_{R,\zeta}, \zeta)$ is minimized with respect to $\zeta \in \mathbb{R}$ and $\tilde{u}_{R,\zeta,0}$ in $H^1(\Omega_R)$. We note that (4.5) is crucial for the iteration process, since otherwise the update $\tilde{u}_k$ in Algorithm 4.1 would not depend on the previous $\zeta_k$.

**Algorithm 4.1.** Minimization of the energy $\mathcal{E}(\tilde{u}_{R,\zeta}, \zeta)$.

**Step 0:** Pick any $\tilde{u}_{R,1}, \tilde{u}_{R,2} \in H^1(\Omega_R)$ with (4.4) and (4.5) and set $k := 1$ and $\zeta_k := 1$.

**Step 1:** Minimize the quadratic energy

$\mathcal{E}_k(u) := \mathcal{E}\left(u + \tilde{u}_{R,1} + \frac{\zeta_k}{R} \tilde{u}_{R,2}, \zeta_k\right)$

with respect to $u \in \overset{\circ}{H}^1(\Omega_R)$, i.e., find $u \in \overset{\circ}{H}^1(\Omega_R)$ such that

$\forall \varphi \in \overset{\circ}{H}^1(\Omega_R)$

$\langle \nabla u, \nabla \varphi \rangle_{L^2(\Omega_R)} = -\langle \nabla \tilde{u}_{R,1}, \nabla \varphi \rangle_{L^2(\Omega_R)} - \frac{\zeta_k}{R} \langle \nabla \tilde{u}_{R,2}, \nabla \varphi \rangle_{L^2(\Omega_R)}$.

Set $\tilde{u}_k := u$. 

Step 2: Minimize the second-order polynomial

\[ p_k(\zeta) := \mathcal{E}(\bar{u}_k + \bar{u}_{R,1} + \frac{\zeta}{R} \bar{u}_{R,2}, \zeta) = \left| \nabla \bar{u}_k + \nabla \bar{u}_{R,1} + \frac{\zeta}{R} \nabla \bar{u}_{R,2} \right|_{L^2(\Omega_R)}^2 + \frac{4\pi \zeta^2}{R} \]

with respect to \( \zeta \), i.e., set

\[ \zeta := -R \frac{\langle \nabla(\bar{u}_k + \bar{u}_{R,1}), \nabla \bar{u}_{R,2} \rangle_{L^2(\Omega_R)}}{4\pi R + |\nabla \bar{u}_{R,2}|_{L^2(\Omega_R)}^2}. \]

Set \( \zeta_{k+1} := \zeta \).

Step 3: Set \( k := k + 1 \) and return to Step 1, unless \( |\zeta_k - \zeta_{k-1}|/|\zeta_k| \) is small.

Step 4: Set \( \tilde{u}_{R,0} := \bar{u}_{k-1} \), \( \zeta := \zeta_k \) and

\[ \tilde{u} := \begin{cases} 
\tilde{u}_{R,0} + \bar{u}_{R,1} + \frac{\zeta}{R} \bar{u}_{R,2} & \text{in } \Omega_R \\
\zeta & \text{in } E_R.
\end{cases} \]

The conforming estimates from Theorem 2.1 involving the free variables \( u \in H^1_{-1}(\Omega) \) and \( v \in D(\Omega) \) are used to estimate the approximation error. For the variable \( u \) we simply choose \( u_R \in H^1(\Omega_R) \) and extend \( u_R \) by zero to \( \Omega \), which defines a proper \( u \). To restrict all computations to \( \Omega_R \), the best choice for \( v \) in \( E_R \) is \( v|_{E_R} := \nabla \tilde{u}|_{E_R} = -\zeta r^{-2} e^r \) with the unit radial vector \( e^r(x) := x/|x| \). Picking vector fields \( v_R \) as restrictions from \( D(\Omega) \) to \( \Omega_R \), i.e., \( v_R \in D(\Omega_R) \), we need that the extensions

\[ v := \begin{cases} 
v_R & \text{in } \Omega_R \\
-\zeta r^{-2} e^r & \text{in } E_R
\end{cases} \tag{4.6} \]

belong to \( D(\Omega) \). Hence, \( e^r \cdot v \), the normal component of \( v \), must be continuous across \( S_R \). Therefore, we get on \( S_R \) the transmission condition \( e^r \cdot v_R = -\zeta / R^2 \). Thus, any \( v_R \) in

\[ V(\Omega_R) := \{ \psi \in D(\Omega_R) : e^r \cdot \psi|_{S_R} = -\zeta / R^2 \} \]
which is extended by (4.6) to \( \Omega \), belongs to \( D(\Omega) \). We note that then \( \nabla \tilde{u} = v \) and \( \text{div} v = \Delta \tilde{u} = 0 \) holds in \( E_R \). Now, the estimates from Theorem 2.1 read

\[
|\nabla e|_{L^2(\Omega)}^2 = |\nabla (\hat{u} - \tilde{u})|_{L^2(\Omega)}^2 \\
\leq \inf_{v \in D(\Omega)} \inf_{\beta > 0} \left( 4 \left( 1 + \frac{1}{\beta} \right) |\text{div} v|_{L^2(\Omega)}^2 + (1 + \beta) |\nabla \tilde{u} - v|_{L^2(\Omega)}^2 \right) \\
= \inf_{v \in D(\Omega)} \inf_{\beta > 0} \left( 4 \left( 1 + \frac{1}{\beta} \right) \int_\Omega (1 + r^2) |\text{div} v|^2 dl + (1 + \beta) \int_\Omega |\nabla \tilde{u} - v|^2 dl \right) \\
\leq \inf_{v_R \in \mathcal{V}(\Omega_R)} \inf_{\beta > 0} \left( 4 \left( 1 + \beta \right) \int_{\Omega_R} (1 + r^2) |\text{div} v_R|^2 dl + (1 + \beta) \int_{\Omega_R} |\nabla \tilde{u} - v_R|^2 dl \right) \\
= M_{+R,\beta}^2(\nabla \tilde{u}, v_R)
\]

and

\[
|\nabla e|_{L^2(\Omega)}^2 = |\nabla (\hat{u} - \tilde{u})|_{L^2(\Omega)}^2 \\
\geq \sup_{u \in \tilde{H}^1(\Omega)} - \langle \nabla (2\tilde{u} + u), \nabla u \rangle_{L^2(\Omega)} = \sup_{u \in \tilde{H}^1(\Omega)} - \int_\Omega \nabla (2\tilde{u} + u) \cdot \nabla u dl \\
\geq \sup_{u_R \in \tilde{H}^1(\Omega_R)} - \int_{\Omega_R} \nabla (2\tilde{u} + u_R) \cdot \nabla u_R dl \\
= M_{-R}(\nabla \tilde{u}, u_R)
\]

Therefore, we have reduced the computations of the lower and upper bounds to minimization problems taking place only in the bounded domain \( \Omega_R \).

**Example 4.1.** We set \( \Omega := E_1 \), i.e., \( \Omega \) is the exterior of a closed unit ball. As already mentioned, this problem admits the unique solution \( \hat{u} = 1/r \). Hence, we know \( \zeta = 1 \) a priori, but ignore it for the computations. We use the symmetry of the problem and thus our computations take place in just one octant of \( \Omega_R \). The mesh and the computational domain are depicted in Fig. 2. The resulting relative errors and the error estimates are presented in Table 1 for three different values of \( R \). Additionally, we study the error indicator generated by the majorant (see Remark 2.2). The exact error contribution over an element \( T \) is \( |\nabla e|_{L^2(T)}^2 \) and the one indicated by the second term of the majorant is \( |\nabla \tilde{u} - v_R|_{L^2(T)}^2 \), where \( v_R \) is obtained via the minimization of the majorant. Both quantities are depicted on the plane \( x_1 + x_3 = 0 \) in Fig. 3.
Figure 2. Computational domains and meshes (one octant of $\Omega_R$, where $R = 10$) for Example 4.1 (a) and Example 4.2 (b).

Figure 3. Example 4.1, true error $|\nabla e|^2_{L^2(T)}$ (a) and error indicator $|\nabla \tilde{u} - v_R|^2_{L^2(T)}$ (b).

The boundary value problem in Step 1 of Algorithm 4.1 was solved by the finite element method. We applied first-order nodal tetrahedral elements, where the mesh was constructed by Comsol 4.3 and the emerging system of linear equations was solved using a standard Matlab solver. When minimizing the majorant $M^2_{+,R,B}(\nabla \tilde{u}, v_R)$ with respect to $v_R \in V(\Omega_R)$, we applied second-order tetrahedral finite elements for each component of $v_R$ with $v_R|_{S_R} = -\zeta R^{-2}e^R$ and hence $e^R \cdot v_R|_{S_R} = -\zeta / R^2$. Similarly, the minorant $M^-_{-,R}(\nabla \tilde{u}, u_R)$ was maximized using second-order tetrahedral finite elements for $u_R \in H^1(\Omega_R)$. A natural choice is to use $H(\text{div})$-conforming Raviart–Thomas elements [6] to compute $v_R$. However, $H^1$-elements are applicable for smooth problems.
Table 1. Example 4.1, the exact error and the computed error estimates.

| $R$ | $N_t$ | $M_{\mathcal{E},R}^2(\nabla \tilde{u}, v_R)$ | $|\nabla \tilde{u}|^2_{L^2(\Omega)}$ | $M_{-,R}(\nabla \tilde{u}, u_R)$ |
|-----|-------|----------------------------------|---------------------------------|---------------------------------|
| 5   | 1926  | 4.57%                            | 3.75%                           | 3.61%                           |
| 5   | 6223  | 1.39%                            | 1.31%                           | 1.29%                           |
| 5   | 17742 | 0.63%                            | 0.62%                           | 0.61%                           |
| 10  | 2214  | 7.19%                            | 5.68%                           | 5.43%                           |
| 10  | 5264  | 2.82%                            | 2.55%                           | 2.51%                           |
| 10  | 11506 | 1.80%                            | 1.68%                           | 1.65%                           |
| 20  | 3446  | 7.26%                            | 5.73%                           | 5.47%                           |
| 20  | 6676  | 2.95%                            | 2.64%                           | 2.60%                           |
| 20  | 13370 | 1.86%                            | 1.74%                           | 1.71%                           |

Table 2. Error bounds for Example 4.2.

| $N_t$ | $M_{\mathcal{E},R}^2(\nabla \tilde{u}, v_R)$ | $|\nabla \tilde{u}|^2_{L^2(\Omega)}$ | $M_{-,R}(\nabla \tilde{u}, u_R)$ |
|-------|----------------------------------|---------------------------------|---------------------------------|
| 2513  | 12.52%                           | 4.65%                           |                                 |
| 5015  | 9.48%                            | 2.91%                           |                                 |
| 13772 | 7.08%                            | 1.63%                           |                                 |

**Example 4.2.** We set $\Omega := \mathbb{R}^3 \setminus [-1,1]^3$, i.e., $\Omega$ is the exterior of a closed cube. For this problem, the exact solution $\hat{u}$ is not known. The octant of $\Omega_R$ for $R = 10$ used in the computations is depicted in Fig. 2. In Algorithm 4.1, we have selected $\tilde{u}_{R,1}$ and $\tilde{u}_{R,2}$ as finite element solutions of Dirichlet Laplace problems with proper boundary conditions. The respective error bounds are presented in Table 2.

These examples show that functional *a posteriori* error estimates provide two-sided bounds of the error. Of course, the accuracy depends on the method used to generate the free variables $v$ and $u$ in the majorant and minorant, respectively. The applied methods should be selected balancing the desired accuracy of the error estimate and the computational expenditures.

**Appendix A. Easier but weaker estimates**

We want to point out that we can prove a variant of Theorem 3.1 by another, much simpler technique using just the triangle inequality instead of
the Helmholtz decomposition. The drawbacks are that we get for the upper bound a factor larger than 1, e.g., 5, which overestimates a bit more, and for the lower bound we miss one term. To see this, let $u \in \mathring{H}^{1}_{-1}(\Omega)+\{u_{0}\}$. Then

$$
|E|_{L^{2}(\Omega),A}^{2} = |\nabla \tilde{u} - A^{-1} \tilde{v}|_{L^{2}(\Omega),A}^{2} \leq |\nabla(\tilde{u} - u)|_{L^{2}(\Omega),A} + \nabla|u - A^{-1} \tilde{v}|_{L^{2}(\Omega),A}
$$

and we can further estimate by Theorem 2.1 with $\tilde{u} = u$ for any $v \in D(\Omega)$

$$
|E|_{L^{2}(\Omega),A}^{2} \leq c_{N,\alpha} |f + \text{div} \, v|_{L^{2}_{\Omega}(\Omega)}^{2} + \nabla|u - A^{-1} v|_{L^{2}(\Omega),A}^{2} + 2 \nabla|u - A^{-1} \tilde{v}|_{L^{2}(\Omega),A}^{2}.
$$

Therefore, we get the same upper bound, but with less good factors, i.e., for any $\theta > 0$

$$
|E|_{L^{2}(\Omega),A}^{2} \leq A_+((\tilde{v}) := (1 + \frac{4}{\theta}) \inf_{v \in D(\Omega)} M_{+}^{2}(A^{-1} \tilde{v}, v) + (4 + \theta) \inf_{u \in \mathring{H}^{1}_{-1}(\Omega)+\{u_{0}\}} M_{+}^{2}(A^{-1} \tilde{v}, \nabla u)
$$

with $\tilde{M}_{+}(\tilde{v}) \geq M_{+}(\tilde{v})$. For example, for $\theta = 1$ we have $1 + 4/\theta = 4 + \theta = 5$ and $\tilde{M}_{+}(\tilde{v}) = 5 M_{+}(\tilde{v})$. For the lower bound and $u \in \mathring{H}^{1}_{-1}(\Omega)$ we simply have

$$
|E|_{L^{2}(\Omega),A}^{2} = |\nabla \tilde{u} - A^{-1} \tilde{v}|_{L^{2}(\Omega),A}^{2} \\
\geq 2 \langle \nabla \tilde{u} - A^{-1} \tilde{v}, \nabla u \rangle_{L^{2}(\Omega),A} - \nabla|u|_{L^{2}(\Omega),A}^{2} \\
= 2 \langle f, u \rangle_{L^{2}(\Omega)} - \langle \nabla u + 2A^{-1} \tilde{v}, \nabla u \rangle_{L^{2}(\Omega),A} = M_{-}(A^{-1} \tilde{v}, u)
$$

and the additional term $\tilde{M}_{-}(A^{-1} \tilde{v}, A^{-1} v, \nabla u)$ does not appear. Thus,

$$
|E|_{L^{2}(\Omega),A}^{2} \geq M_{-}(A^{-1} \tilde{v}, u)
$$

with $\tilde{M}_{-}(\tilde{v}) \leq M_{-}(\tilde{v})$.

**Acknowledgement**

We heartily thank Sergey Repin for his continuous support and many interesting and enlightening discussions.
References


